Fast Fourier Transform (FFT)

<u>Problem</u>: we need an efficient way to compute the DFT. The answer is the FFT.

Consider a data sequence x = [x(0), x(1), ..., x(N-1)] and its DFT:

$$X(k) = \sum_{n=0}^{N-1} x(n) w_N^{kn}, \quad k = 0, ..., N-1$$

We can always break the summation into two summations: one on <u>even</u> indices (n=0,2,4,...) and one on <u>odd</u> indeces (n=1,3,5,...), as

$$X(k) = \sum_{n \text{ even}} x(n) w_N^{kn} + \sum_{n \text{ odd}} x(n) w_N^{kn}, \quad k = 0, ..., N-1$$

Let us assume that the total number of points N is <u>even</u>, ie N/2 is an integer. Then we can write the DFT as

$$X(k) = \sum_{n \text{ even}} x(n) w_N^{kn} + \sum_{n \text{ odd}} x(n) w_N^{kn}$$

$$= \sum_{m=0}^{\frac{N}{2}-1} x(2m) w_N^{k(2m)} + \sum_{m=0}^{\frac{N}{2}-1} x(2m+1) w_N^{k(2m+1)}$$

$$= \sum_{m=0}^{\frac{N}{2}-1} x(2m) \left(w_N^2\right)^{km} + w_N^k \sum_{m=0}^{\frac{N}{2}-1} x(2m+1) \left(w_N^2\right)^{km}$$

$$W_{N/2}$$

since
$$w_N^2 = (e^{-j2\mathbf{p}/N})^2 = e^{-j2\mathbf{p}/(N/2)} = w_{N/2}$$

The two summations are two distinct DFT's, as we can see below

$$X(k) = \sum_{m=0}^{\frac{N}{2}-1} x(2m) w_{N/2}^{km} + w_N^{k} \sum_{m=0}^{\frac{N}{2}-1} x(2m+1) w_{N/2}^{km}$$
N-point DFT = N/2-point DFT + w_N^{k} N/2-point DFT

$$X_N(k) = X_{N/2}^e(k) + w_N^k X_{N/2}^o(k)$$

for k=0,...N-1, where

$$X^{e}_{N/2} = DFT[x^{even}], x^{even} = [x(0), x(2), ..., x(N-2)];$$

 $X^{o}_{N/2} = DFT[x^{odd}], x^{odd} = [x(1), x(3), ..., x(N-1)].$

The problem is that in the expression

$$X_N(k) = X_{N/2}^e(k) + w_N^k X_{N/2}^o(k)$$

the N-point DFT and the N/2 point DFT's have different lengths, since we define them as

$$X_N(k), \quad k = 0,...,N-1$$

 $X_{N/2}(k), \quad k = 0,...,N/2-1$

For example if N=4 we need to compute

$$X_4 = [X_4(0), X_4(1), X_4(2), X_4(3)]$$

from two 2-point DFT's
$$X^e = [X_2^e(0), X_2^e(1)]$$

 $X^o = [X_2^o(0), X_2^o(1)]$

So how do we compute $X_4(2)$ and $X_4(3)$?

We use the <u>periodicity</u> of the DFT, and relate the N-point DFT with the two N/2-point DFT's as follows

$$X_{N}(k) = X^{e}_{N/2}(k) + w_{N}^{k} X^{o}_{N/2}(k)$$

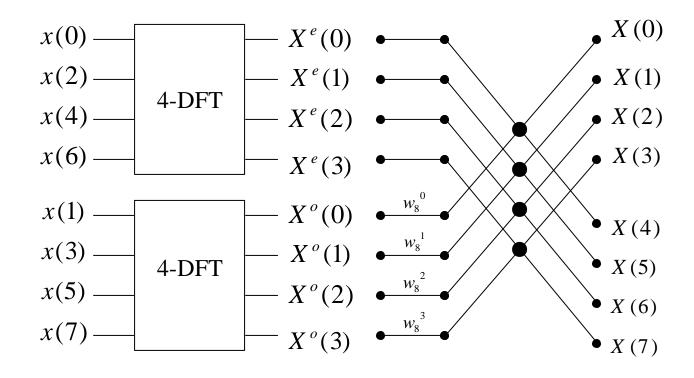
$$X_{N}\left(k + \frac{N}{2}\right) = X^{e}_{N/2}(k) - w_{N}^{k} X^{o}_{N/2}(k), \qquad k = 0, ..., \frac{N}{2} - 1$$

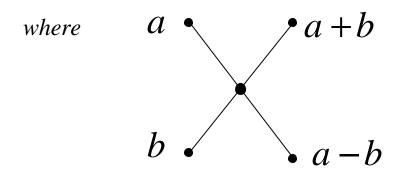
where we used the facts that

• the DFT is periodic and in particular $X_{N/2}(k) = X_{N/2}\left(k + \frac{N}{2}\right);$

•
$$w_N^{k+N/2} = w_N^k e^{-(j2\mathbf{p}/N)N/2} = -w_N^k$$

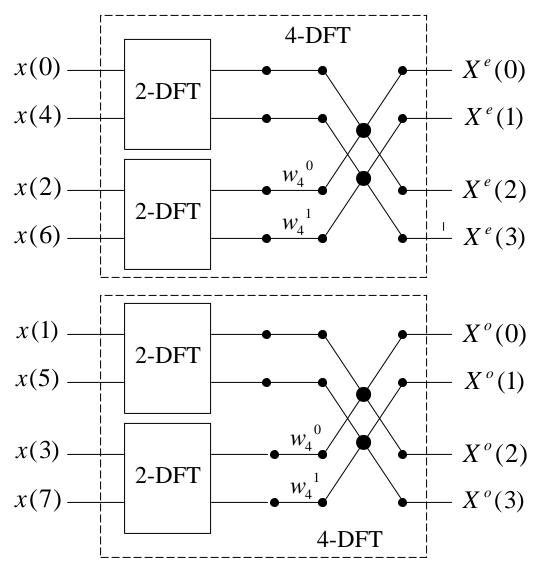
General Structure of the FFT (take, say, N=8):



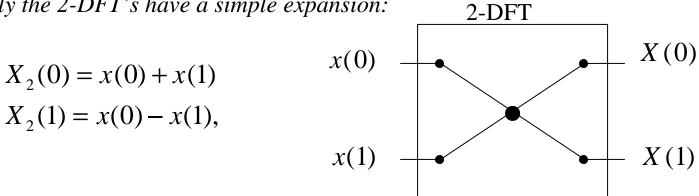


is called the <u>butterfly</u>.

Same for the 4-DFT:



Finally the 2-DFT's have a simple expansion:



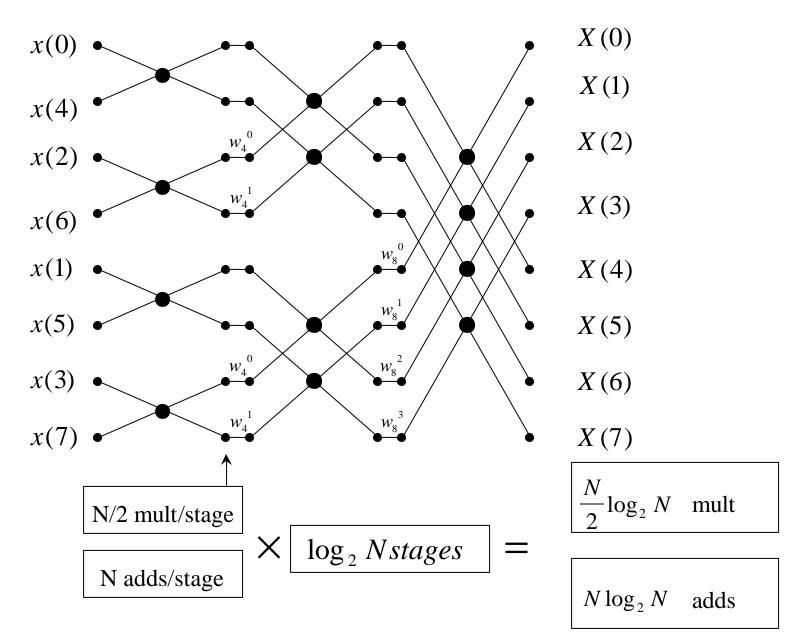
since

$$X_2(k) = \sum_{n=0}^{1} x(n)w_2^{kn} = x(0) + w_2^{k} x(1)$$

with

$$w_2^k = e^{-j\mathbf{p}\,k} = (-1)^k$$

Put everything together:



We say that, for a data set of length $N = 2^{L}$,

complexity of the FFT is
$$O\{N \log_2 N\}$$

i.e. number of operations $\leq a N \log_2 N + b$ for some constants a,b.

On the other hand, for the same data of length N,

complexity of the DFT is
$$O\{N^2\}$$

Since, from the formula,
$$X(k) = \sum_{n=0}^{N-1} x(n) w_N^{kn}, \quad k = 0,..., N-1$$

N ops/term
$$\times$$
 N terms $=$ N² ops

This is a big difference in the total number of computations, as shown in this graph:

complexity

